

A Diagnostic Function for Bootstrap Confidence Intervals

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The Standard Intervals

- Observe $\hat{\theta}$ as estimate of θ , standard error σ
- Approximate 95% two-sided interval for θ

$$\hat{\theta} \pm 1.96\sigma$$

- Exactly accurate and correct in a **normal translation family**

$$\hat{\theta} = \theta + \sigma Z, \quad Z \sim \mathcal{N}(0, 1)$$

- Automatic!

Student Score Data

Mardia and Bibby

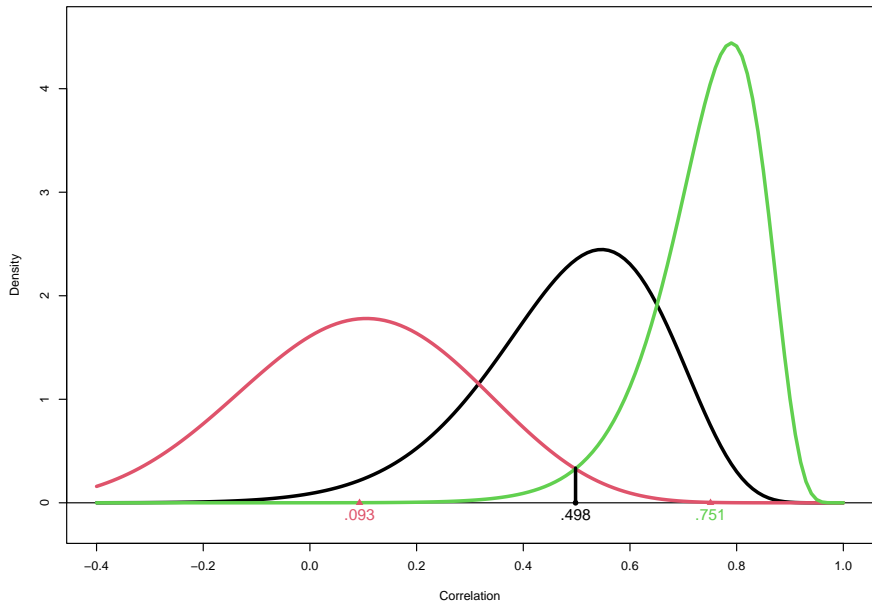
mechanics	vectors	algebra	analytics	statistics
7	51	43	17	22
44	69	53	53	53
49	41	61	49	64
59	70	68	62	56
34	42	50	47	29
⋮	⋮	⋮	⋮	⋮
42	69	61	55	45
46	49	53	59	37
63	63	65	70	63

Case Study: The Student Score Data

Mardia and Bibby

- $n = 22$ students have each taken five tests
- **Normal model** $\mathbf{X}_{22 \times 5}$ has independent rows $\mathbf{x}_i \sim \mathcal{N}_5(\boldsymbol{\mu}, \boldsymbol{\Sigma})$
- *Parameter of interest* $\theta = \text{correlation}(\text{mechanics}, \text{vectors})$
- MLE $\hat{\theta} = 0.496$, $\hat{\sigma} = \frac{(1 - \hat{\theta}^2)}{\sqrt{n - 3}} = 0.173$
- Standard 95% interval $\theta \in (.160, .836)$

Density of $\hat{\theta}$ for $\theta = .498$, and for the exact .95 confidence limits $\theta = .093$ and $\theta = .751$



Fisher's Transformation Trick

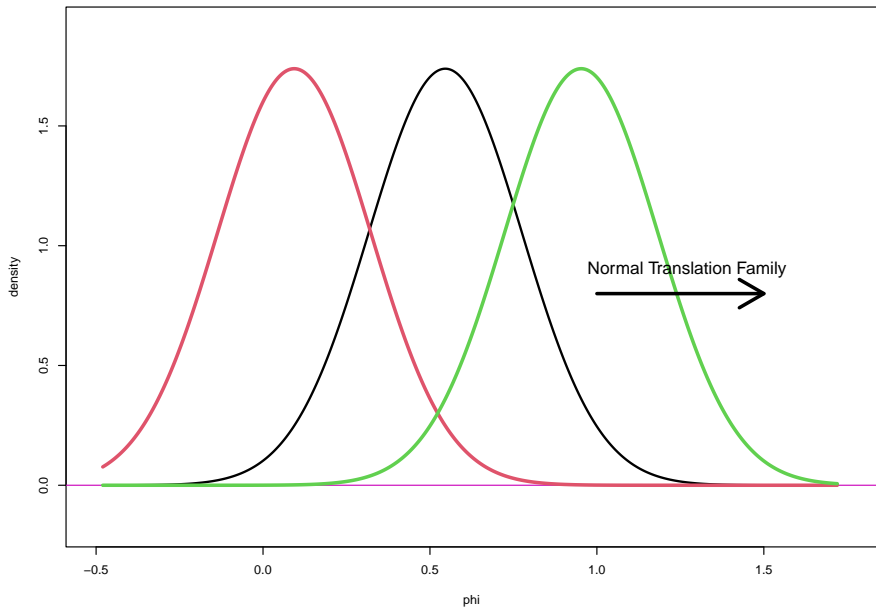
- Monotone transformation $\phi = m(\theta)$, $\hat{\phi} = m(\hat{\theta})$

$$m(\theta) = \frac{1}{2} \log \left(\frac{1 + \theta}{1 - \theta} \right)$$

- **Normal translation family** $\hat{\phi} \sim \mathcal{N}(\phi, \sigma^2)$, $\sigma^2 = (n - 3)^{-1}$
- Apply standard method on ϕ scale: $\hat{\phi} \pm 1.96\sigma$
- *Inverse transformation* $\hat{\theta} = m^{-1}(\hat{\phi})$ back to θ scale:

$$\theta \in m^{-1}(\hat{\phi} \pm 1.96\sigma)$$

The three densities on Fisher's phi scale



Exact & Approximate Endpoints

for 95% interval

	.025	.975
Standard	.160	.836
Fisher	.097	.760
Exact	.093	.751
BCA	.086	.754

The Trouble With Fisher's Method

1. Requires Fisher
 2. No nuisance parameters
- `bca` “bias-corrected and accelerated”
 - A bootstrap algorithm for more accurate approximate confidence intervals
 - automatable — doesn't require a Fisher
 - allows for nuisance parameters

Maximum Eigenvalue Example

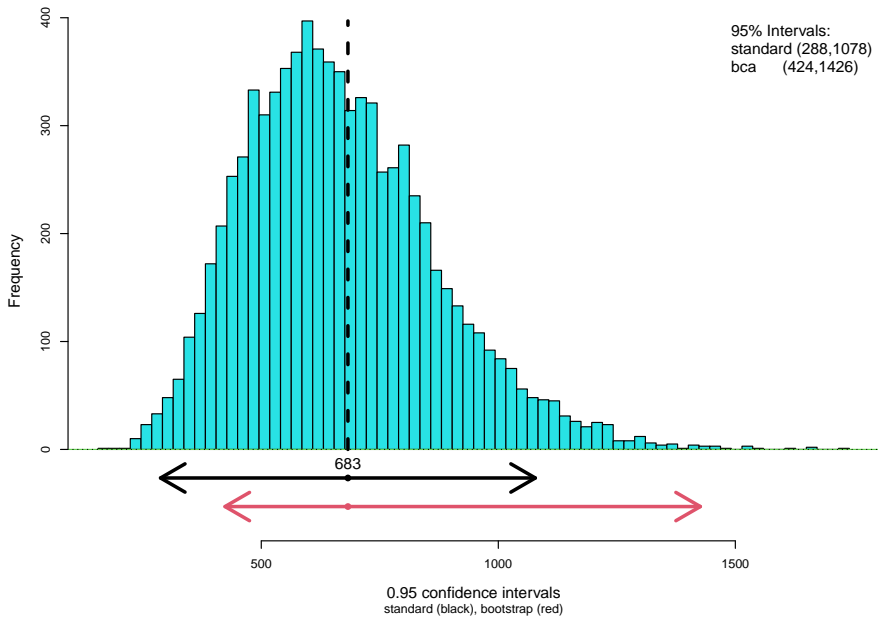
- Student score data matrix $\mathbf{X}_{22 \times 5}$, assuming rows $x_i \stackrel{\text{ind}}{\sim} \mathcal{N}_5(\mu, \Sigma)$
- Now let θ be the maximum eigenvalue of Σ , with MLE $\hat{\theta}$

$$\hat{\theta} = \text{maximum eigenvalue}(\hat{\Sigma}) = 683$$

- $\mathcal{N}_5(\mu, \Sigma)$ is a 20-parameter family so there are 19 nuisance parameters
- **Parametric bootstrap** $x_i^* \stackrel{\text{ind}}{\sim} \mathcal{N}_5(\hat{\mu}, \hat{\Sigma}), i = 1, \dots, 22$

$$\implies \mathbf{X}_{22 \times 5}^* \implies \hat{\Sigma}^* \implies \hat{\theta}^*$$

8000 bootstraps, maximum eigenvalue



Good Things About The BCA Intervals

- Automatable \rightarrow one program handles all cases
- Transformation invariant \rightarrow if $\phi = m(\theta)$ then $\hat{\phi}[\alpha] = m(\hat{\theta}[\alpha])$
($\hat{\theta}[\alpha]$ is bca level α upper endpoint)
- Second-order accurate \rightarrow actual coverage approaches nominal α at $O(n^{-1})$ compared to $O(n^{-1/2})$ for standard
- **Correctness** \rightarrow $\hat{\theta}[\alpha]$ is *correct* as well as accurate

NSTF: Normal Scaled Transformation Families

(Efron, 1982: *Ann Stat* 323–339)

- A one-parameter family of cdfs $\{F_\theta(\hat{\theta})\}$ is **NSTF** if for some monotone transformation $\phi = m(\theta)$ and $\hat{\phi} = m(\hat{\theta})$ we have

$$\hat{\phi} = \phi + \sigma_\phi(Z - z_0) \quad [\sigma_\phi = 1 + a_0\phi, Z \sim \mathcal{N}(0, 1)]$$

- z_0 “bias corrector”
- a_0 “acceleration”
- **Fisher** $a_0 = z_0 = 0$ and $m(\theta) = \frac{1}{2} \log\left(\frac{1+\theta}{1-\theta}\right)$

BCA Confidence Limits

(Efron, 1987: JASA 171–200)

- If $F_{\hat{\theta}}$ is NSTF, the exact and correct one-sided upper level α confidence limit for θ is

$$\hat{\theta}[\alpha] = F_{\hat{\theta}}^{-1} \left\{ \Phi \left(z_0 + \frac{z_0 + z^{(\alpha)}}{1 - a_0(z_0 + z^{(\alpha)})} \right) \right\}$$

- Don't need to know $m(\cdot)$!
- Gives standard interval if $a_0 = z_0 = 0$ and $F_{\hat{\theta}}(\hat{\theta}) = \mathcal{N}(\theta, \sigma^2)$

Is The Family $\{F_\theta(\hat{\theta})\}$ An NSTF?

- For a given value of z let $\alpha = \Phi(z)$,

$$\hat{\theta}^{(\alpha)} = \alpha\text{th quantile of } \hat{\theta} \quad (= F_\theta^{-1}(\alpha))$$

$$\text{and } \dot{F}_\theta(\hat{\theta}) = \frac{\partial}{\partial \theta} F_\theta(\hat{\theta}).$$

- The diagnostic function is

$$D_\theta(z) = \frac{\dot{F}_\theta(\hat{\theta}^{(\alpha)}) / \dot{F}_\theta(\hat{\theta}^{(0.5)})}{\phi(z) / \phi(0)}$$

Theorem

$\{F_\theta(\hat{\theta})\}$ is an NSTF iff $D_\theta(z)$ is linear in z .

Gamma Scale Family

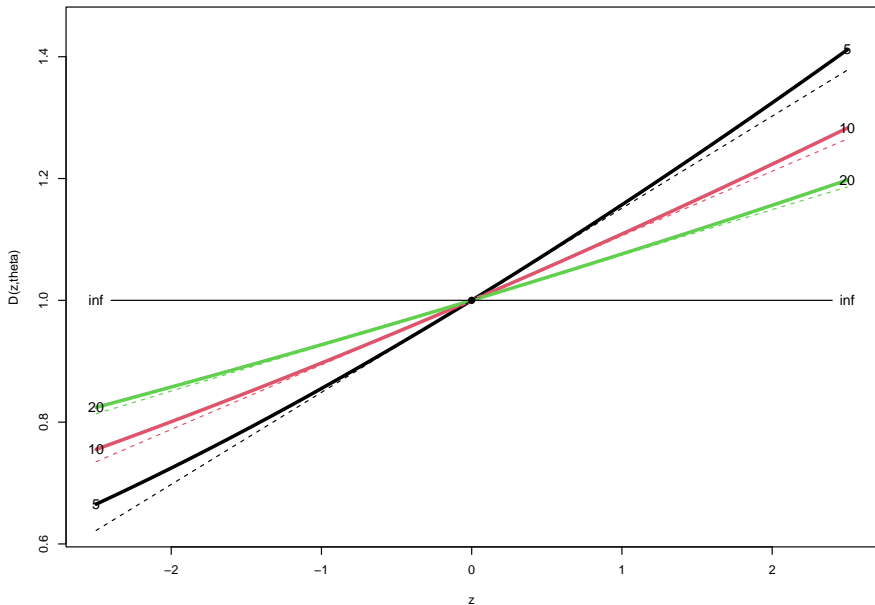
$$\hat{\theta} = \theta \cdot \text{Gamma}_\nu$$

$$f_\theta(\hat{\theta}) = \frac{\hat{\theta}^{\nu-1} e^{-\hat{\theta}/\theta}}{\theta^\nu \Gamma(\nu)} \quad \text{for } \hat{\theta} \geq 0$$

$$F_\theta(\hat{\theta}) = \int_0^{\hat{\theta}} f_\theta(t) dt$$

- Next $D_\theta(z)$ (same for all θ)

Diagnostic function $D(z)$ for Gamma scale family,
df=5, 10, 20 and Infinity. Dashed lines are linear fits



Student- t Translation Family

- One-parameter family

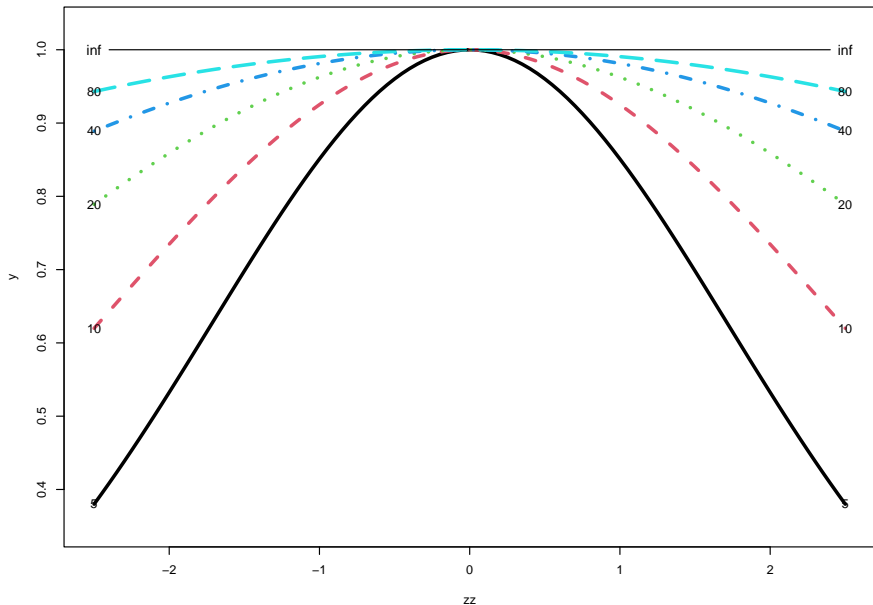
$$\hat{\theta} = \theta + W$$

where

W Student's t distribution with ν degrees of freedom

- Next $D_{\theta}(z)$ (same for all θ)

Diagnostic Function for Student-t translation model;
df = 5, 10, 20, 40, 80 and Infinity



Normal Theory Correlation Coefficient

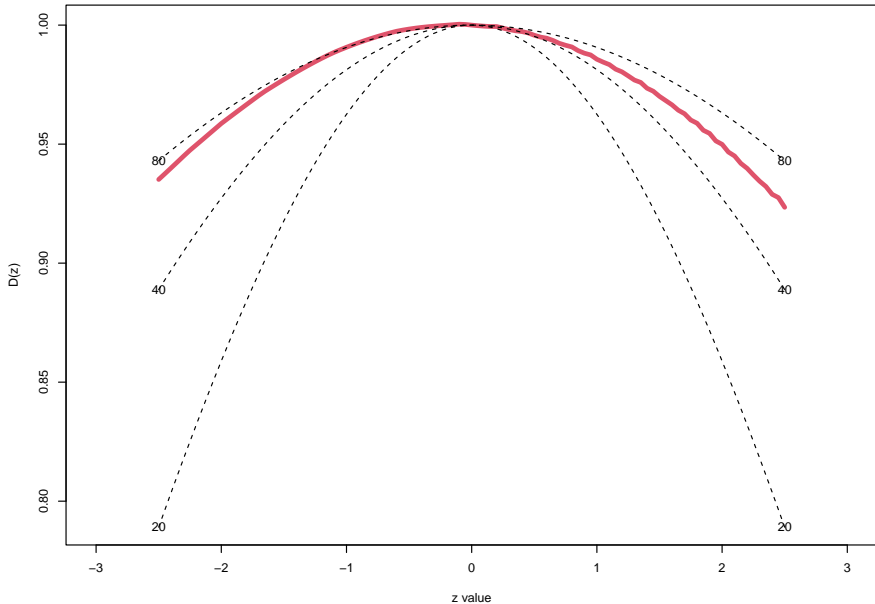
$$x_i \stackrel{\text{ind}}{\sim} \mathcal{N}_5(\mu, \Sigma), \quad i = 1, \dots, 22$$

$\theta = \text{correlation}(X_1, X_2)$

$\hat{\theta} = \text{Pearson correlation coefficient}$

- Next $D_{.498}(z)$

Diagnostic function $D(z)$ for normal correlation, $\theta = .498$;
 $n = 22$; Compared with student-t translation models



GSTF: General Scaled Transformation Family

- **NSTF** $\hat{\phi} = \phi + (1 + a_0\phi)(Z - z_0)$ with $Z \sim \mathcal{N}(0, 1)$
- **GSTF** Replace Z with $W = w(Z)$
- $w(0) = 0$ and $w'(0) = 1$
- *Example* $W \sim \text{student}_\nu$ so $w(z) = T_\nu^{-1}(\Phi(z))$
- If $a_0 = z_0 = 0$: $\hat{\phi} = \phi + W$

Diagnostic Function For A GSTF

Theorem

If $F_{\theta}(\hat{\theta})$ is a GSTF then

$$D_{\theta}(z) = \frac{1 + w(z)a_0}{w'(z)}.$$

- For NSTF, $w(z) = z$, $w'(z) = 1$ and, for every value of θ ,

$$D_{\theta}(z) = 1 + za_0,$$

linear in z , with slope a_0 .

GSTF Confidence Limits

- **Fact:** For any choice of $w(\cdot)$ there's a version of the bca limits that is accurate and correct for that GSTF.
- *Idea:* Use estimated diagnostic function to estimate “ w ” and see if the GSTF limits are much different than the NSTF bca limit.

Multiparameter Models & Nuisance Parameters

- *Exponential families*: Observe data \mathbf{X} with densities

$$f_{\eta}(\mathbf{X}) \propto e^{\eta^{\top} \mathbf{y}} \quad (\eta, \mathbf{y} \in \mathcal{R}^p)$$

- $\mathbf{y} = \mathbf{y}(\mathbf{X})$ sufficient vector and $\eta =$ natural parameter vector
- $\mu = E_{\eta}\{\mathbf{y}\}$ “expectation parameter”
- Parameter of interest

$$\theta = t(\mu), \quad \text{with } \hat{\theta} = t(\mathbf{y}) \text{ MLE}$$

- $p - 1$ nuisance parameters

Stein's Least Favorable Family

- Replace p -parameter family $f_{\eta}(\mathbf{X})$ with one-parameter family

$$\hat{f}_{\lambda}(\mathbf{X}) = f_{\hat{\eta} + \lambda t}(\mathbf{X}) \begin{cases} \hat{\eta} & = \text{MLE } \eta, \text{ fixed} \\ t & = (\dots \partial t(\mathbf{y}) / \partial y_i \dots) \end{cases}$$

- Back to one-parameter model
- Apply bca formula, diagnostic function, ...

Student Score Maximum Eigenvalue Example

- Data $\mathbf{X}_{22 \times 5}$ has rows $x_i \stackrel{\text{ind}}{\sim} \mathcal{N}_5(\mu, \Sigma)$, $i = 1, \dots, 22$
- $y = (\bar{X}_1, \bar{X}_2, \dots, \bar{X}_5, \overline{X_1^2}, \overline{X_1 X_2}, \dots, \overline{X_5^2})$
- $p = 20$
- $\theta = \text{max eigenvalue}(\Sigma)$ has MLE $\hat{\theta} = \text{maxeig}(\widehat{\text{cov}}(\mathbf{X}))$

Calculating The BCA Intervals

(Efron and Narasimhan, 2020: CRAN package `bcaboot`)

- **Parametric bootstrap** $\hat{\eta}$ MLE of η ,

$$f_{\hat{\eta}}(\cdot) \longrightarrow \mathbf{X}_1^*, \mathbf{X}_2^*, \dots, \mathbf{X}_B^*$$

and

$$y_i^* = y(\mathbf{X}_i^*), \quad \hat{\theta}_i^* = t(y_i^*) \quad \text{for } i = 1, \dots, B$$

- Program `bcapar` computes NSTF bca confidence limits; nonparametric version is `bcanon`
- Also gives diagnostic function $\widehat{D}(z)$: *Linear?*

Estimated Diagnostic Function

- First-order Taylor approximation to $\hat{\theta}_i^*$ is

$$\hat{\theta}_i^* = \hat{\theta} + d_i^* \quad \text{where}$$

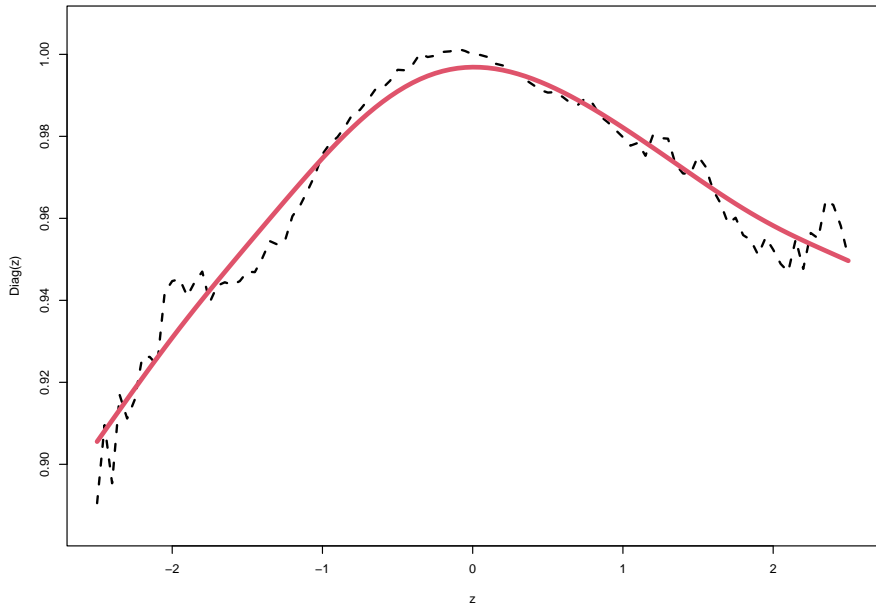
$$d_i^* = t^\top (y_i^* - \bar{y}^*)$$

Lemma

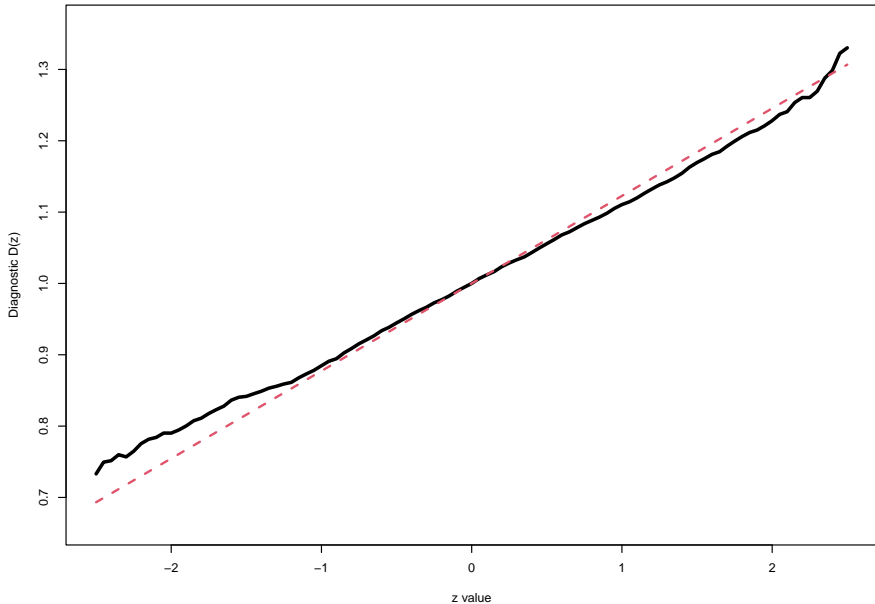
$$\left. \frac{\partial}{\partial \lambda} \hat{F}_\lambda(z) \right|_{\lambda=0} = \frac{1}{B} \sum_{\hat{\theta}_i^* \leq \hat{\theta}^*(\alpha)} d_i^*$$

- $\alpha = \Phi(z)$ and $\hat{\theta}^*(\alpha) = \text{boot } \alpha\text{th quantile}$
- Gives diagnostic estimate $\hat{D}_0(z)$

Estimated and Smoothed Diagnostic Function for the Student Score Correlation



Diagnostic Function Student Score Maximum Eigenvalue (and linear fit)



Actual GSTF Levels

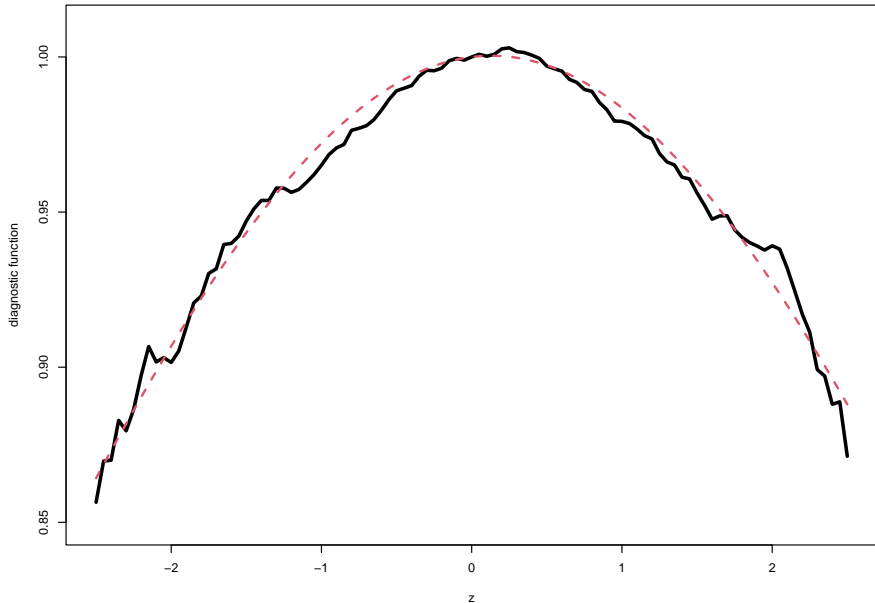
for nominal NSTF endpoints

	Student score maximum eigenvalue			
Nominal NSTF	.025	.050	.950	.975
Actual GSTF	.024	.049	.948	.975

Nonparametric BCA Confidence Intervals

- Student score data matrix $\mathbf{X}_{22 \times 5}$ gives $\hat{\theta} = t(\mathbf{X})$
- **Bootstrap replication** $\mathbf{X}^*_{22 \times 5}$: rows selected from \mathbf{X} with replacement gives $\hat{\theta}^* = t(\mathbf{X}^*)$
- `bcanon(B,X,func)` gives bca confidence limits and diagnostic
- *Next* $B = 8000$ `func = cor(X1, X2)`

Nonparametric Diagnostic Function estimate, correlation example;
Red is cubic fit; Close to student-t translation model $df=44$



Truth and Targets

